



# Classifying Out-Of-Vocabulary Terms in a Domain-Specific Social Media Corpus

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## Introduction

- High rate of out-of-vocabulary (OOV) terms in social media text.
- Presents challenge to most natural language processing (NLP) systems as they rely heavily on lexical knowledge.
- Goal: automatically classify OOV terms in automotive web forums into domain specific categories.
- Coarse-grained categories could serve as a preliminary source of lexical knowledge about the out-of-vocabulary terms

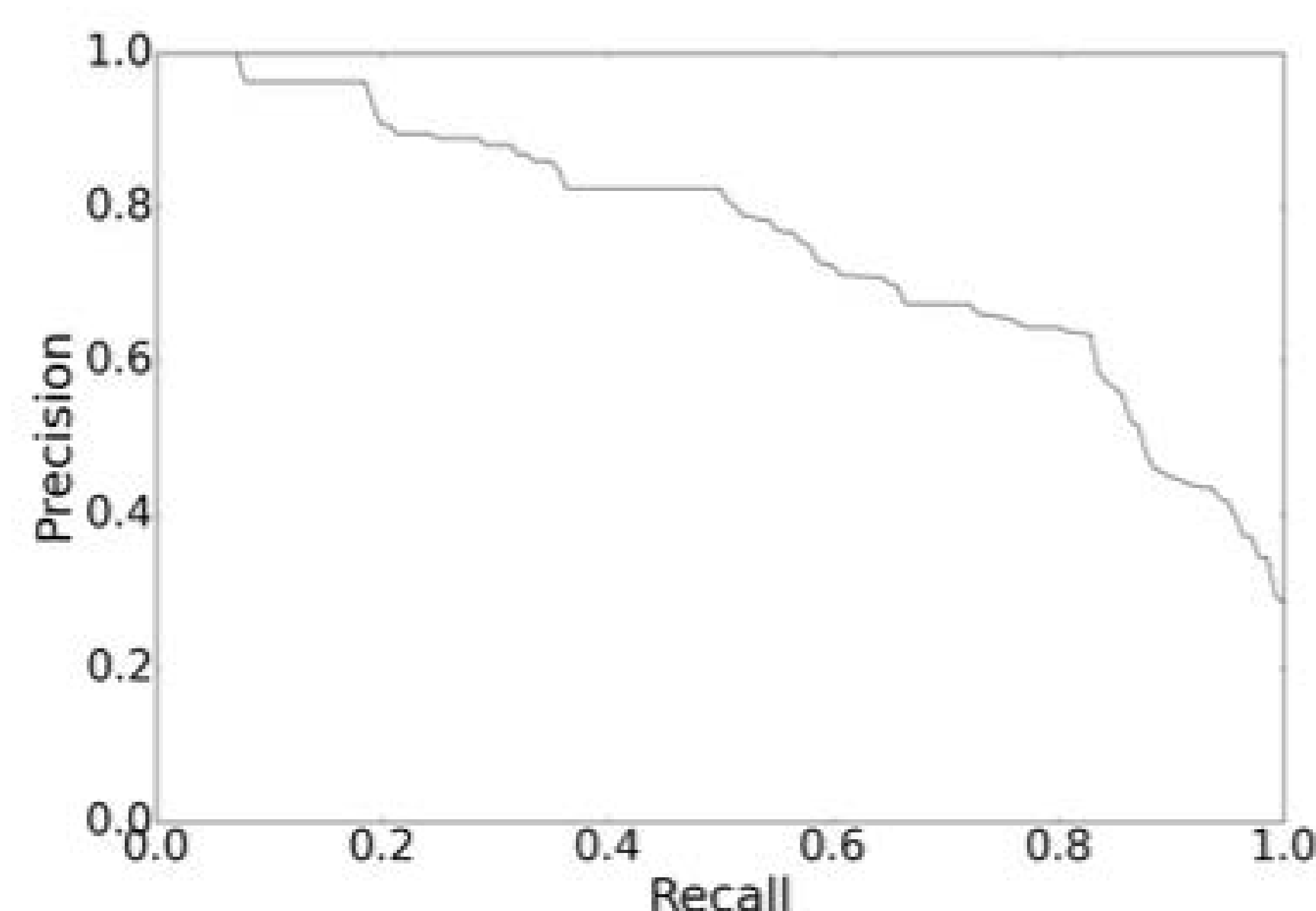
## Methods

- Supervised learning approach
- Features sets:
  - Character N-grams
  - Language models
  - Frequency
  - Word embedding
  - Surface form
- Experimental setup: 10x10-fold cross-validation logistic regression

Category	Num. items	Explanation	Examples
AUTO	45	Automotive terms (not NEs)	<i>defuel, rebalance</i>
DRUG	95	Drug names	<i>levoxyl, nexium</i>
FOREIGN	47	Non-English terms	<i>rezeptfrei, depuis</i>
MEASUREMENT	58	Units of measurement	<i>77k, 100mph</i>
NE-AUTO	140	Automotive-related NEs	<i>ls3, volks</i>
NE-OTHER	41	Non-automotive NEs	<i>blackhawks, diaz</i>
NOISE	87	Noise, and items that don't fit other categories	<i>kagvjfcjfx, kzvddzfv52</i>
SLANG	59	Internet slang and non-standard forms	<i>heyyaa, lol2</i>
SPELLING-ERROR	93	Spelling errors	<i>youll, genericfor</i>

## Results

Method	Precision	Recall	F1 score	Accuracy
Most-frequent class baseline	0.023	0.111	0.039	0.211
[A] Character <i>n</i> -grams (1-3)	0.390	0.373	0.380	0.413
[B] Language models	0.023	0.111	0.039	0.211
[C] Frequency	0.023	0.111	0.039	0.211
[D] Word embeddings	0.649	0.599	0.622	0.643
[E] Surface form	0.390	0.400	0.394	0.446
[A+B+C+D+E]	0.643	0.603	0.622	0.649
[B+C+D+E]	0.649	0.602	0.624	0.646
[A+C+D+E]	0.640	0.605	0.622	0.648
[A+B+D+E]	<b>0.650</b>	<b>0.609</b>	<b>0.628</b>	<b>0.654</b>
[A+B+C+E]	0.429	0.422	0.424	0.469
[A+B+C+D]	0.614	0.582	0.597	0.629



- Interpolated precision-recall curve for ranking based on probability of NE-AUTO class
- Ranking can be useful for semi-automatic identification of NE-AUTO terms

- Word embedding features are very informative of OOV meaning